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**ECON 351\* -- NOTE 23**
**Tests for Coefficient Differences: Examples 2**
**1. Introduction**
**Model and Data**

- **Sample data:** A random sample of 534 paid employees.
- **Variable definitions:**

$W_i$  = hourly wage rate of employee  $i$ ;

$\ln W_i$  = the natural logarithm of  $W_i$ ;

$S_i$  = years of schooling completed by employee  $i$ ;

$X_i$  = years of work experience accumulated by employee  $i$ .

$F_i$  = a female indicator variable, = 1 if employee  $i$  is female, 0 otherwise;

$M_i$  = a male indicator variable, = 1 if employee  $i$  is male, 0 otherwise.

- **The Model:** A simple log-lin (semi-log) wage equation of the form

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + u_i \quad (1)$$

- **Two Groups of Employees: Female and Male**
- ◆ The *female* wage equation

$$\ln W_i = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i + u_i \quad i = 1, \dots, N_f = 245 \quad (2.1)$$

- ◆ The *male* wage equation

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + u_i \quad i = 1, \dots, N_m = 289 \quad (2.2)$$

## 2. Tests for Full Coefficient Equality

### Null and Alternative Hypotheses

$$H_0: \alpha_0 = \beta_0 \text{ and } \alpha_1 = \beta_1 \text{ and } \alpha_2 = \beta_2 \quad (3 \text{ coefficient restrictions})$$

$$H_1: \alpha_0 \neq \beta_0 \text{ and/or } \alpha_1 \neq \beta_1 \text{ and/or } \alpha_2 \neq \beta_2.$$

### Unrestricted Model – Approach 1: Separate Female and Male Wage Equations

#### ♦ The *female* wage equation

$$\ln W_i = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i + u_i \quad i = 1, \dots, N_f = 245 \quad (2.1)$$

The OLS sample regression equation for *females* (with t-ratios) is:

$$\begin{aligned} \ln W_i &= \hat{\alpha}_0 + \hat{\alpha}_1 S_i + \hat{\alpha}_2 X_i + \hat{u}_i & i = 1, \dots, N_f = 245 & \quad K_0 = 3 \\ \hat{\alpha}_0 &= 0.3031 & \hat{\alpha}_1 = 0.1117 & \quad \hat{\alpha}_2 = 0.008854 & \quad \mathbf{RSS}_{(1)} = 43.2866 \\ & (1.699) & (9.381) & \quad (3.868) & \quad \mathbf{df}_{(1)} = 245 - 3 = 242 \end{aligned}$$

#### ♦ The *male* wage equation

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + u_i \quad i = 1, \dots, N_m = 289 \quad (2.2)$$

The OLS sample regression equation for *males* (with t-ratios) is:

$$\begin{aligned} \ln W_i &= \hat{\beta}_0 + \hat{\beta}_1 S_i + \hat{\beta}_2 X_i + \hat{u}_i & i = 1, \dots, N_m = 289 & \quad K_0 = 3 \\ \hat{\beta}_0 &= 0.6965 & \hat{\beta}_1 = 0.09030 & \quad \hat{\beta}_2 = 0.01635 & \quad \mathbf{RSS}_{(2)} = 63.1474 \\ & (4.279) & (8.430) & \quad (6.690) & \quad \mathbf{df}_{(2)} = 289 - 3 = 286 \end{aligned}$$

#### ♦ The *unrestricted* residual sum-of-squares is:

$$\mathbf{RSS}_1 = \mathbf{RSS}_{(1)} + \mathbf{RSS}_{(2)} = 43.2866 + 63.1474 = \mathbf{106.4340};$$

$$\mathbf{df}_1 = \mathbf{df}_{(1)} + \mathbf{df}_{(2)} = 242 + 286 = \mathbf{528}.$$

**Unrestricted Model – Approach 2: Pooled Full-Interaction Wage Equations**

- ◆ **No base group** – no intercept coefficient (t-ratios in parentheses)

$$\ln W_i = \alpha_0 F_i + \alpha_1 F_i S_i + \alpha_2 F_i X_i + \beta_0 M_i + \beta_1 M_i S_i + \beta_2 M_i X_i + u_i \quad (12.0)$$

$$i = 1, \dots, N = N_f + N_m = 534$$

$$\text{RSS}_1 = 106.4340 \quad \text{with} \quad \text{df}_1 = N - 2K_0 = 534 - 6 = 528$$

$$\begin{array}{lll} \hat{\alpha}_0 = 0.3031 & \hat{\alpha}_1 = 0.1117 & \hat{\alpha}_2 = 0.008854 \\ (1.601) & (8.837) & (3.643) \\ \hat{\beta}_0 = 0.6965 & \hat{\beta}_1 = 0.09030 & \hat{\beta}_2 = 0.01635 \\ (4.478) & (8.822) & (7.002) \end{array}$$

$$\begin{array}{lll} \hat{\alpha}_0 - \hat{\beta}_0 = -0.3933 & \hat{\alpha}_1 - \hat{\beta}_1 = 0.02144 & \hat{\alpha}_2 - \hat{\beta}_2 = -0.007491 \\ (-1.605) & (1.318) & (-2.223) \end{array}$$

$$\begin{array}{lll} \hat{\beta}_0 - \hat{\alpha}_0 = 0.3933 & \hat{\beta}_1 - \hat{\alpha}_1 = -0.02144 & \hat{\beta}_2 - \hat{\alpha}_2 = 0.007491 \\ (1.605) & (-1.318) & (2.223) \end{array}$$

Compare coefficient estimates of pooled regression equation (12.0) with the *separate* female and male sample wage equations:

- The OLS sample regression equation for *females* is:

$$\ln W_i = \hat{\alpha}_0 + \hat{\alpha}_1 S_i + \hat{\alpha}_2 X_i + \hat{u}_i \quad i = 1, \dots, N_f = 245 \quad K_0 = 3$$

$$\begin{array}{llll} \hat{\alpha}_0 = 0.3031 & \hat{\alpha}_1 = 0.1117 & \hat{\alpha}_2 = 0.008854 & \text{RSS}_{(1)} = 43.2866 \\ (1.699) & (9.381) & (3.868) & \text{df}_{(1)} = 245 - 3 = 242 \end{array}$$

- The OLS sample regression equation for *males* is:

$$\ln W_i = \hat{\beta}_0 + \hat{\beta}_1 S_i + \hat{\beta}_2 X_i + \hat{u}_i \quad i = 1, \dots, N_m = 289 \quad K_0 = 3$$

$$\begin{array}{llll} \hat{\beta}_0 = 0.6965 & \hat{\beta}_1 = 0.09030 & \hat{\beta}_2 = 0.01635 & \text{RSS}_{(2)} = 63.1474 \\ (4.279) & (8.430) & (6.690) & \text{df}_{(2)} = 289 - 3 = 286 \end{array}$$

◆ **Females as base group** (t-ratios in parentheses)

$$\ln W_i = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i + \gamma_0 M_i + \gamma_1 M_i S_i + \gamma_2 M_i X_i + u_i \quad (12.1)$$

where  $\gamma_0 = \beta_0 - \alpha_0$ ;  $\gamma_1 = \beta_1 - \alpha_1$ ;  $\gamma_2 = \beta_2 - \alpha_2$ .

**RSS<sub>1</sub> = 106.4340** with **df<sub>1</sub> = N - 2K<sub>0</sub> = 534 - 6 = 528**

$\hat{\alpha}_0 = 0.3031$ (1.601)	$\hat{\alpha}_1 = 0.1117$ (8.837)	$\hat{\alpha}_2 = 0.008854$ (3.643)
$\hat{\gamma}_0 = 0.3933$ (1.605)	$\hat{\gamma}_1 = -0.02144$ (-1.318)	$\hat{\gamma}_2 = 0.007491$ (2.223)

From OLS-SRE (12.0):

$\hat{\beta}_0 - \hat{\alpha}_0 = 0.3933$ (1.605)	$\hat{\beta}_1 - \hat{\alpha}_1 = -0.02144$ (-1.318)	$\hat{\beta}_2 - \hat{\alpha}_2 = 0.007491$ (2.223)
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◆ **Males as base group** (t-ratios in parentheses)

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + \delta_0 F_i + \delta_1 F_i S_i + \delta_2 F_i X_i + u_i \quad (12.2)$$

where  $\delta_0 = \alpha_0 - \beta_0$ ;  $\delta_1 = \alpha_1 - \beta_1$ ;  $\delta_2 = \alpha_2 - \beta_2$ .

**RSS<sub>1</sub> = 106.4340** with **df<sub>1</sub> = N - 2K<sub>0</sub> = 534 - 6 = 528**

$\hat{\beta}_0 = 0.6965$ (4.478)	$\hat{\beta}_1 = 0.09030$ (8.822)	$\hat{\beta}_2 = 0.01635$ (7.002)
$\hat{\delta}_0 = -0.3933$ (-1.605)	$\hat{\delta}_1 = 0.02144$ (1.318)	$\hat{\delta}_2 = -0.007491$ (-2.223)

From OLS-SRE (12.0):

$\hat{\alpha}_0 - \hat{\beta}_0 = -0.3933$ (-1.605)	$\hat{\alpha}_1 - \hat{\beta}_1 = 0.02144$ (1.318)	$\hat{\alpha}_2 - \hat{\beta}_2 = -0.007491$ (-2.223)
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**Restricted Model – Same for Approach 1 and Approach 2**

Corresponds to the null hypothesis that all female and male coefficients are equal:

$$H_0: \alpha_0 = \beta_0 \text{ and } \alpha_1 = \beta_1 \text{ and } \alpha_2 = \beta_2 \quad \text{in pooled equation (12.0)}$$

$$\gamma_0 = 0 \text{ and } \gamma_1 = 0 \text{ and } \gamma_2 = 0 \quad \text{in pooled equation (12.1)}$$

$$\delta_0 = 0 \text{ and } \delta_1 = 0 \text{ and } \delta_2 = 0 \quad \text{in pooled equation (12.2)}$$

- ◆ The **restricted model** can be written as *either*

$$\ln W_i = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i + u_i \quad i = 1, \dots, N = 534 \quad (1.1)$$

or

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + u_i \quad i = 1, \dots, N = 534 \quad (1.2)$$

- ◆ The **restricted OLS-SRE** for the full sample of 534 observations is

$$\ln W_i = \tilde{\beta}_0 + \tilde{\beta}_1 S_i + \tilde{\beta}_2 X_i + \tilde{u}_i, \quad i = 1, \dots, N = N_1 + N_2 = 534$$

$$\begin{array}{ccc} \tilde{\beta}_0 = 0.5828 & \tilde{\beta}_1 = 0.09642 & \tilde{\beta}_2 = 0.01175 \\ (4.646) & (11.601) & (6.700) \end{array}$$

- ◆ The **restricted residual sum-of-squares** is:

$$RSS_0 = 117.0626 \quad \text{with} \quad df_0 = N - K_0 = 534 - 3 = 531.$$

### The F-Test for Equality of All Coefficients Between Males and Females

$$F = \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} = \frac{(RSS_0 - RSS_1)/(K - K_0)}{RSS_1/(N - K)} \sim F[df_0 - df_1, df_1]$$

$$\begin{aligned} RSS_0 &= 117.0626 & \text{with} & & df_0 &= N - K_0 = 534 - 3 = 531 \\ RSS_1 &= 106.4340 & \text{with} & & df_1 &= N - 2K_0 = 534 - 6 = 528 \end{aligned}$$

- **Sample value of the F-statistic:**

$$\begin{aligned} F_0 &= \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} \\ &= \frac{(117.0626 - 106.4340)/(531 - 528)}{106.4340/528} \quad \Rightarrow \quad \mathbf{F_0 = 17.58} \\ &= \frac{10.6286/3}{106.4340/528} \\ &= 17.58 \end{aligned}$$

- **Null distribution of  $F_0$ :**  $F_0 \sim F[K_0, N - 2K_0] = F[3, 528]$  under  $H_0$ .

- **Decision Rule:** At the  $100\alpha$  percent significance level

1. **reject  $H_0$**  if  $F_0 \geq F_\alpha[K_0, N - 2K_0]$  or p-value for  $F_0 \leq \alpha$ ;
2. **retain  $H_0$**  if  $F_0 < F_\alpha[K_0, N - 2K_0]$  or p-value for  $F_0 > \alpha$ .

- **Critical Values of  $F[3, 528]$ :** at the 5% and 1% significance levels.

$$\text{At } \alpha = 0.05: \quad F_{0.05}[3, 528] = 2.622$$

$$\text{At } \alpha = 0.01: \quad F_{0.01}[3, 528] = 3.819$$

- **P-value for  $F_0 = 0.0000$ .**

- **Inference:**

Since  $F_0 = 17.58 > 3.819 = F_{0.01}[3, 528]$ , **reject  $H_0$**  at the **1%** significance level.

Since **p-value for  $F_0 = 0.0000 < 0.01$** , **reject  $H_0$**  at the **1%** significance level.

### 3. Tests for Equality of Both Slope Coefficients

#### Null and Alternative Hypotheses

$$\text{female } S_i \text{ coefficient} = \text{male } S_i \text{ coefficient} \quad \Rightarrow \quad \alpha_1 = \beta_1$$

$$\text{female } X_i \text{ coefficient} = \text{male } X_i \text{ coefficient} \quad \Rightarrow \quad \alpha_2 = \beta_2$$

$$H_0: \alpha_1 = \beta_1 \text{ and } \alpha_2 = \beta_2 \quad \text{in pooled equation (12.0)}$$

$$\gamma_1 = 0 \text{ and } \gamma_2 = 0 \quad \text{in pooled equation (12.1)}$$

$$\delta_1 = 0 \text{ and } \delta_2 = 0 \quad \text{in pooled equation (12.2)}$$

against

$$H_1: \alpha_1 \neq \beta_1 \text{ and/or } \alpha_2 \neq \beta_2 \quad \text{in pooled equation (12.0)}$$

$$\gamma_1 \neq 0 \text{ and/or } \gamma_2 \neq 0 \quad \text{in pooled equation (12.1)}$$

$$\delta_1 \neq 0 \text{ and/or } \delta_2 \neq 0 \quad \text{in pooled equation (12.2)}$$

#### Unrestricted Model – corresponds to $H_1$

Any one of the three pooled full-interaction regression equations for  $\ln W_i$ .

$$\ln W_i = \alpha_0 F_i + \alpha_1 F_i S_i + \alpha_2 F_i X_i + \beta_0 M_i + \beta_1 M_i S_i + \beta_2 M_i X_i + u_i \quad (12.0)$$

$$i = 1, \dots, N = N_f + N_m = 534$$

$$\ln W_i = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i + \gamma_0 M_i + \gamma_1 M_i S_i + \gamma_2 M_i X_i + u_i \quad (12.1)$$

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + \delta_0 F_i + \delta_1 F_i S_i + \delta_2 F_i X_i + u_i \quad (12.2)$$

- OLS estimation of any one of the three unrestricted regression equations (12.0), (12.1) or (12.2) on the pooled sample of 534 male and female employees yields the following value for the *unrestricted* RSS:

$$RSS_1 = 106.4340 \quad \text{with} \quad df_1 = N - 2K_0 = 534 - 6 = 528.$$

**Restricted Model** – corresponds to  $H_0$ 

Obtained by substituting the two coefficient restrictions specified by the null hypothesis  $H_0$  into any one of the three pooled full-interaction regression equations for  $\ln W_i$ .

- ♦ **In equation (12.0) – no base group: set  $\alpha_1 = \beta_1$  and  $\alpha_2 = \beta_2$**

$$\ln W_i = \alpha_0 F_i + \alpha_1 F_i S_i + \alpha_2 F_i X_i + \beta_0 M_i + \beta_1 M_i S_i + \beta_2 M_i X_i + u_i \quad (12.0)$$

$$\begin{aligned} \ln W_i &= \alpha_0 F_i + \beta_1 F_i S_i + \beta_2 F_i X_i + \beta_0 M_i + \beta_1 M_i S_i + \beta_2 M_i X_i + u_i \\ &= \alpha_0 F_i + \beta_1 (F_i S_i + M_i S_i) + \beta_2 (F_i X_i + M_i X_i) + \beta_0 M_i + u_i \\ &= \alpha_0 F_i + \beta_0 M_i + \beta_1 (F_i + M_i) S_i + \beta_2 (F_i + M_i) X_i + u_i \\ &= \alpha_0 F_i + \beta_0 M_i + \beta_1 S_i + \beta_2 X_i + u_i \quad \text{since } F_i + M_i = 1 \quad \forall i. \end{aligned}$$

$$\ln W_i = \alpha_0 F_i + \beta_0 M_i + \beta_1 S_i + \beta_2 X_i + u_i \quad (13.0)$$

- ♦ **In equation (12.1) – females as base group: set  $\gamma_1 = 0$  and  $\gamma_2 = 0$**

$$\ln W_i = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i + \gamma_0 M_i + \gamma_1 M_i S_i + \gamma_2 M_i X_i + u_i \quad (12.1)$$

$$\ln W_i = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i + \gamma_0 M_i + u_i \quad (13.1)$$

- ♦ **In equation (12.2) – males as base group: set  $\delta_1 = 0$  and  $\delta_2 = 0$**

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + \delta_0 F_i + \delta_1 F_i S_i + \delta_2 F_i X_i + u_i \quad (12.2)$$

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + \delta_0 F_i + u_i \quad (13.2)$$

- OLS estimation of any one of the three restricted regression equations (13.0), (13.1) or (13.2) on the pooled sample of 534 male and female employees yields the following value for the **restricted RSS**:

$$\mathbf{RSS}_0 = 108.4501 \quad \text{with} \quad \mathbf{df}_0 = \mathbf{N} - \mathbf{K}_0 = 534 - 4 = 530.$$

### The F-Test for Male-Female Equality of *Both* Slope Coefficients

$$F = \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} = \frac{(RSS_0 - RSS_1)/(K - K_0)}{RSS_1/(N - K)} \sim F[df_0 - df_1, df_1]$$

$$RSS_0 = 108.4501 \quad \text{with} \quad df_0 = N - K_0 = 534 - 4 = 530$$

$$RSS_1 = 106.4340 \quad \text{with} \quad df_1 = N - K = 534 - 6 = 528$$

- **Sample value of the F-statistic:**

$$\begin{aligned} F_0 &= \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} \\ &= \frac{(108.4501 - 106.4340)/(530 - 528)}{106.4340/528} \quad \Rightarrow \quad \mathbf{F_0 = 5.001} \\ &= \frac{2.0161/2}{106.4340/528} \\ &= 5.001 \end{aligned}$$

- **Null distribution of  $F_0$ :**  $F_0 \sim F[df_0 - df_1, df_1] = F[2, 528]$  under  $H_0$ .

- **Decision Rule:** At the  $100\alpha$  percent significance level

1. **reject  $H_0$**  if  $F_0 \geq F_\alpha[df_0 - df_1, df_1]$  or p-value for  $F_0 \leq \alpha$ ;

2. **retain  $H_0$**  if  $F_0 < F_\alpha[df_0 - df_1, df_1]$  or p-value for  $F_0 > \alpha$ .

- **Critical Values of  $F[2, 528]$ :** at the 5% and 1% significance levels.

$$\text{At } \alpha = 0.05: \quad F_{0.05}[2, 528] = 3.013$$

$$\text{At } \alpha = 0.01: \quad F_{0.01}[2, 528] = 4.646$$

- **P-value for  $F_0 = 0.00706$ .**

- **Inference:**

Since  $F_0 = 5.001 > 4.646 = F_{0.01}[2, 528]$ , **reject  $H_0$**  at the **1%** significance level.

Since **p-value for  $F_0 = 0.00706 < 0.01$** , **reject  $H_0$**  at the **1%** significance level.

#### 4. Conditional Log-Wage Differentials Between Males and Females

**Question:** What is the **mean log-wage differential** between **male and female employees** with the *same* education and work experience – i.e., with the *same* values of **S** and **X**?

□ **Conditional *female-male* mean log-wage differential** is:

$$E(\ln W | S_i, X_i, F_i = 1) - E(\ln W | S_i, X_i, F_i = 0)$$

or

$$E(\ln W | S_i, X_i, M_i = 0) - E(\ln W | S_i, X_i, M_i = 1).$$

◆ The ***female* log-wage equation** is

$$\ln W_i = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i + u_i \quad i = 1, \dots, N_f = 245 \quad (2.1)$$

Hence, the **conditional mean log-wage of *females* with  $S_i$  years of completed schooling and  $X_i$  years of work experience** is:

$$E(\ln W | S_i, X_i, F_i = 1) = E(\ln W | S_i, X_i, M_i = 0) = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i \quad (14.1)$$

◆ The ***male* log-wage equation** is

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + u_i \quad i = 1, \dots, N_m = 289 \quad (2.2)$$

Hence, the **conditional mean log-wage of *males* with  $S_i$  years of completed schooling and  $X_i$  years of work experience** is:

$$E(\ln W | S_i, X_i, F_i = 0) = E(\ln W | S_i, X_i, M_i = 1) = \beta_0 + \beta_1 S_i + \beta_2 X_i \quad (14.2)$$

- ◆ The **conditional female-male mean log-wage differential** is obtained by **subtracting (14.2) from (14.1)**:

$$E(\ln W | S_i, X_i, F_i = 1) = E(\ln W | S_i, X_i, M_i = 0) = \alpha_0 + \alpha_1 S_i + \alpha_2 X_i \quad (14.1)$$

$$E(\ln W | S_i, X_i, F_i = 0) = E(\ln W | S_i, X_i, M_i = 1) = \beta_0 + \beta_1 S_i + \beta_2 X_i \quad (14.2)$$

$$\begin{aligned} E(\ln W | S_i, X_i, F_i = 1) - E(\ln W | S_i, X_i, F_i = 0) \\ &= \alpha_0 + \alpha_1 S_i + \alpha_2 X_i - \beta_0 - \beta_1 S_i - \beta_2 X_i \\ &= (\alpha_0 - \beta_0) + (\alpha_1 - \beta_1) S_i + (\alpha_2 - \beta_2) X_i \\ &= \delta_0 + \delta_1 S_i + \delta_2 X_i \end{aligned}$$

$$\text{where } \delta_0 = \alpha_0 - \beta_0; \quad \delta_1 = \alpha_1 - \beta_1; \quad \delta_2 = \alpha_2 - \beta_2.$$

- ◆ **Pooled full interaction log-wage equation with *males* as base group**

$$\ln W_i = \beta_0 + \beta_1 S_i + \beta_2 X_i + \delta_0 F_i + \delta_1 F_i S_i + \delta_2 F_i X_i + u_i \quad (12.2)$$

$$\text{where } \delta_0 = \alpha_0 - \beta_0; \quad \delta_1 = \alpha_1 - \beta_1; \quad \delta_2 = \alpha_2 - \beta_2.$$

- **OLS estimation of pooled full interaction log-wage equation (12.2) yields the sample regression equation**

$$\ln W_i = \hat{\beta}_0 + \hat{\beta}_1 S_i + \hat{\beta}_2 X_i + \hat{\delta}_0 F_i + \hat{\delta}_1 F_i S_i + \hat{\delta}_2 F_i X_i + \hat{u}_i \quad (12.2^*)$$

The **OLS coefficient estimates (and t-ratios)** for (12.2) are:

$$\begin{array}{lll} \hat{\beta}_0 = 0.6965 & \hat{\beta}_1 = 0.09030 & \hat{\beta}_2 = 0.01635 \\ (4.478) & (8.822) & (7.002) \\ \hat{\delta}_0 = -0.3933 & \hat{\delta}_1 = 0.02144 & \hat{\delta}_2 = -0.007491 \\ (-1.605) & (1.318) & (-2.223) \end{array}$$

- ◆ **The *estimate* of the conditional *female-male* mean log-wage differential is**

$$\begin{aligned}
 & \hat{E}(\ln W | S_i, X_i, F_i = 1) - \hat{E}(\ln W | S_i, X_i, F_i = 0) \\
 &= (\hat{\alpha}_0 - \hat{\beta}_0) + (\hat{\alpha}_1 - \hat{\beta}_1)S_i + (\hat{\alpha}_2 - \hat{\beta}_2)X_i \\
 &= \hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i \\
 &= -0.3933 + 0.02144S_i - 0.007491X_i
 \end{aligned} \tag{15}$$

**Example:** The estimated **conditional *female-male* mean log-wage differential** for employees with **16 years of schooling and 10 years of work experience** is obtained by setting  $S_i = 16$  and  $X_i = 10$  in equation (15):

$$\begin{aligned}
 & \hat{E}(\ln W | S_i = 16, X_i = 10, F_i = 1) - \hat{E}(\ln W | S_i = 16, X_i = 10, F_i = 0) \\
 &= -0.3933 + 0.02144S_i - 0.007491X_i \\
 &= -0.3933 + 0.02144(16) - 0.007491(10) = -0.12517
 \end{aligned}$$

**Interpretation:** The **average wage of *female* employees** with 16 years of schooling and 10 years of work experience is **approximately 12.5 percent less** than the **average wage of *male* employees** with the same schooling and work experience.

- **The *variance* of the conditional *female-male* mean log-wage differential is**

$$\begin{aligned}
 \text{Var}(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i) &= \text{Var}(\hat{\delta}_0) + S_i^2 \text{Var}(\hat{\delta}_1) + X_i^2 \text{Var}(\hat{\delta}_2) \\
 &\quad + 2S_i \text{Cov}(\hat{\delta}_0, \hat{\delta}_1) + 2X_i \text{Cov}(\hat{\delta}_0, \hat{\delta}_2) + 2S_i X_i \text{Cov}(\hat{\delta}_1, \hat{\delta}_2)
 \end{aligned}$$

**The *standard error* of the conditional *female-male* mean log-wage differential** is simply the ***square root* of the variance**:

$$\text{se}(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i) = \sqrt{\text{Var}(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i)}$$

***Proposition:*** The conditional female-male mean log-wage differential for employees with 16 years of schooling and 10 years of work experience equals zero.

**Null and Alternative Hypotheses**

$$H_0: E(\ln W | S_i = 16, X_i = 10, F_i = 1) - E(\ln W | S_i = 16, X_i = 10, F_i = 0) = 0$$

$$\text{or } \delta_0 + \delta_1 S_i + \delta_2 X_i = \delta_0 + 16\delta_1 + 10\delta_2 = 0$$

$$H_1: E(\ln W | S_i = 16, X_i = 10, F_i = 1) - E(\ln W | S_i = 16, X_i = 10, F_i = 0) \neq 0$$

$$\text{or } \delta_0 + \delta_1 S_i + \delta_2 X_i = \delta_0 + 16\delta_1 + 10\delta_2 \neq 0$$

**Perform a two-tail t-test:** The *t*-statistic is

$$t(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i) = \frac{\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i - (\delta_0 + \delta_1 S_i + \delta_2 X_i)}{se(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i)} \sim t[N - 2K_0] \quad (16)$$

- Calculate the *estimated variance and standard error* of

$$\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i = \hat{\delta}_0 + 16\hat{\delta}_1 + 10\hat{\delta}_2:$$

$$\begin{aligned} \text{Var}(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i) &= \text{Var}(\hat{\delta}_0) + S_i^2 \text{Var}(\hat{\delta}_1) + X_i^2 \text{Var}(\hat{\delta}_2) \\ &\quad + 2S_i \text{Cov}(\hat{\delta}_0, \hat{\delta}_1) + 2X_i \text{Cov}(\hat{\delta}_0, \hat{\delta}_2) + 2S_i X_i \text{Cov}(\hat{\delta}_1, \hat{\delta}_2) \end{aligned}$$

```
. regress lnw s x f fs fx
```

Source	SS	df	MS	Number of obs =	534
Model	42.0079479	5	8.40158958	F( 5, 528) =	41.68
Residual	106.433962	528	.201579473	Prob > F =	0.0000
				R-squared =	0.2830
				Adj R-squared =	0.2762
Total	148.44191	533	.278502645	Root MSE =	.44898

lnw	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
s	.0903006	.0102354	8.822	0.000	.0701936 .1104076
x	.0163455	.0023344	7.002	0.000	.0117597 .0209314
f	-.3933286	.2450724	-1.605	0.109	-.8747652 .0881081
fs	.0214448	.0162689	1.318	0.188	-.0105149 .0534046
fx	-.0074912	.0033699	-2.223	0.027	-.0141112 -.0008712
_cons	.6964687	.155538	4.478	0.000	.3909194 1.002018

```

. matrix list VC2

symmetric VC2[6,6]
      s          x          f          fs          fx          _cons
s    .00010476
x    8.540e-06    5.449e-06
f    .00151678    .00020904    .06006048
fs   -.00010476   -8.540e-06   -.00381452   .00026468
fx   -8.540e-06   -5.449e-06   -.00046734   .00001939
_cons -.00151678  -.00020904  -.02419208   .00151678   .00020904   .02419208

```

$$\text{Vâr}(\hat{\delta}_0) = 0.06006048$$

$$\text{Vâr}(\hat{\delta}_1) = 0.00026468$$

$$\text{Vâr}(\hat{\delta}_2) = 0.00001136$$

$$\text{Côv}(\hat{\delta}_0, \hat{\delta}_1) = -0.00381452$$

$$\text{Côv}(\hat{\delta}_0, \hat{\delta}_2) = -0.00046734$$

$$\text{Côv}(\hat{\delta}_1, \hat{\delta}_2) = 0.00001939$$

- **Set  $S_i = 16$  and  $X_i = 10$**  in formula for  $\text{Vâr}(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i)$ :

$$\begin{aligned} \text{Vâr}(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i) &= \text{Vâr}(\hat{\delta}_0) + S_i^2 \text{Vâr}(\hat{\delta}_1) + X_i^2 \text{Vâr}(\hat{\delta}_2) \\ &\quad + 2S_i \text{Côv}(\hat{\delta}_0, \hat{\delta}_1) + 2X_i \text{Côv}(\hat{\delta}_0, \hat{\delta}_2) + 2S_i X_i \text{Côv}(\hat{\delta}_1, \hat{\delta}_2) \end{aligned}$$

- **Calculate *estimated variance* of  $\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i = \hat{\delta}_0 + 16\hat{\delta}_1 + 10\hat{\delta}_2$ :**

$$\begin{aligned} \text{Vâr}(\hat{\delta}_0 + 16\hat{\delta}_1 + 10\hat{\delta}_2) &= \text{Vâr}(\hat{\delta}_0) + (16)^2 \text{Vâr}(\hat{\delta}_1) + (10)^2 \text{Vâr}(\hat{\delta}_2) \\ &\quad + 2(16)\text{Côv}(\hat{\delta}_0, \hat{\delta}_1) + 2(10)\text{Côv}(\hat{\delta}_0, \hat{\delta}_2) + 2(16)(10)\text{Côv}(\hat{\delta}_1, \hat{\delta}_2) \\ &= 0.06006048 + (16)^2(0.00026468) + (10)^2(0.00001136) \\ &\quad + 2(16)(-0.00381452) + 2(10)(-0.00046734) + 2(16)(10)(0.00001939) \end{aligned}$$

### Results:

$$\text{Vâr}(\hat{\delta}_0 + 16\hat{\delta}_1 + 10\hat{\delta}_2) = \mathbf{0.00374538}$$

$$\text{sê}(\hat{\delta}_0 + 16\hat{\delta}_1 + 10\hat{\delta}_2) = \sqrt{\text{Vâr}(\hat{\delta}_0 + 16\hat{\delta}_1 + 10\hat{\delta}_2)} = \sqrt{0.00374538} = \mathbf{0.0611995}$$

- The *sample value of the t-statistic* under  $H_0$  is calculated from (16) by setting

$$\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i = \hat{\delta}_0 + 16\hat{\delta}_1 + 10\hat{\delta}_2 = -\mathbf{0.12512354}$$

$$\delta_0 + \delta_1 S_i + \delta_2 X_i = \delta_0 + 16\delta_1 + 10\delta_2 = \mathbf{0}$$

$$s\hat{e}(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i) = s\hat{e}(\hat{\delta}_0 + 16\hat{\delta}_1 + 10\hat{\delta}_2) = \mathbf{0.0611995}$$

$$t(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i) = \frac{\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i - (\delta_0 + \delta_1 S_i + \delta_2 X_i)}{s\hat{e}(\hat{\delta}_0 + \hat{\delta}_1 S_i + \hat{\delta}_2 X_i)} \quad (16)$$

$$t_0(\hat{\delta}_0 + 16\hat{\delta}_1 + 10\hat{\delta}_2) = \frac{-0.12512354 - 0}{0.0611995} = \frac{-0.12512354}{0.0611995} = -\mathbf{2.0445}$$

- Null distribution of  $t_0$ :**  $t_0 \sim t[N - 2K_0] = t[534 - 2(3)] = t[534 - 6] = t[528]$

- Two-tail critical values of  $t[528]$ :** at the 5% and 1% significance levels

$$\alpha = 0.05 \Rightarrow \alpha/2 = 0.025: \quad t_{0.025}[528] = 1.964$$

$$\alpha = 0.01 \Rightarrow \alpha/2 = 0.005: \quad t_{0.005}[528] = 2.585$$

- Two-tail p-value for  $t_0 = 0.04140$ .**

- Inference:**

Since  $|t_0| = 2.045 < 2.585 = t_{0.005}[528]$ , *retain  $H_0$*  at the **1% significance level**.

Since  $|t_0| = 2.045 > 1.964 = t_{0.025}[528]$ , *reject  $H_0$*  at the **5% significance level**.

Since **p-value for  $t_0 = 0.04140 > 0.01$** , *retain  $H_0$*  at the **1% significance level**.

Since **p-value for  $t_0 = 0.04140 < 0.05$** , *reject  $H_0$*  at the **5% significance level**.

- How to use *Stata* to compute this t-test: Use the following *lincom* command.

```
. lincom _b[f] + 16*_b[fs] + 10*_b[fx]
```

```
( 1)  f + 16.0 fs + 10.0 fx = 0.0
```

	lnw	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
(1)		-.1251235	.0611995	-2.045	0.041	-.245348 - .0048991